

Agenda

- 1. Depopulation in Illinois
- 2. Impact of COVID-19
- 3. Integration with other infrastructure

Depopulation

Depopulation

Objectives

- 1. Identify and cluster depopulating cities (completed)
- 2. Survey depopulating cities (in progress)
- 3. Forecast impact of depopulation in Illinois (in progress)
- 4. Organize a forum at UIC, inviting officials from cities that are facing depopulation, to discuss challenges and solutions (future)
- 5. Sharing all information and models developed openly (future)

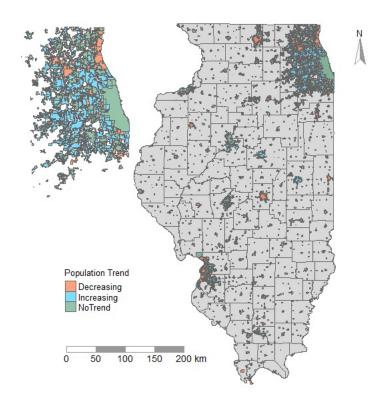
Identify and Cluster Depopulating Cities

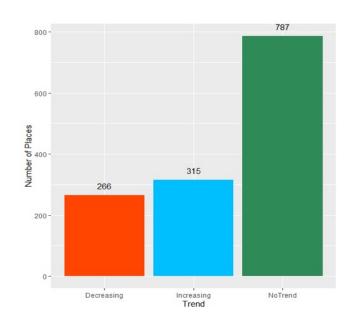
Identify Depopulating Cities

- Depopulation is defined as "chronic population loss."
- Using ACS data, we collected population data from 2009 to 2017 and used Mann-Kendall (MK) test.
- Focus on "places" as defined by the Census Bureau.
 - Smaller than counties but larger than zip codes.
 - Includes municipalities, villages, and boroughs.
 - -> refer to them as "cities" from now on.

Identify Depopulating Cities

Out of 1,368 places in Illinois, 266 were found to be depopulating.





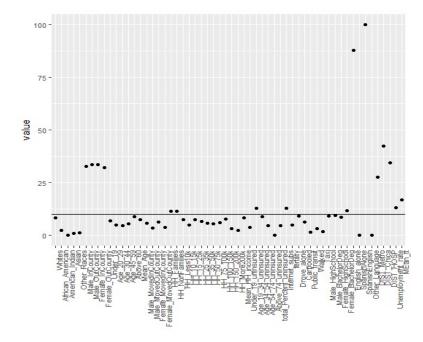
- Manually select variables that seem relevant from Census (not relevant to depopulation but to clustering cities)
 - -> Found 58 variables.

| Dimension | Variables | Categories: Values are in %. |
|-----------|------------------------|---|
| | Language | English only Spanish only Spanish and English Other languages |
| | Household Structure | Families, Non-Families. |
| Social | Uninsured | Age: Under 19, Age: 19 – 34, Age: 35 – 54, Age: 55 – 74, Age: Above 75. |
| | | Uninsured percentage |
| | Employment | Male: Work in same county, Male: Work in different county, Female: Work in same county, Female: Work in different |
| | | county. |
| | Unemployment | Unemployment percentage |

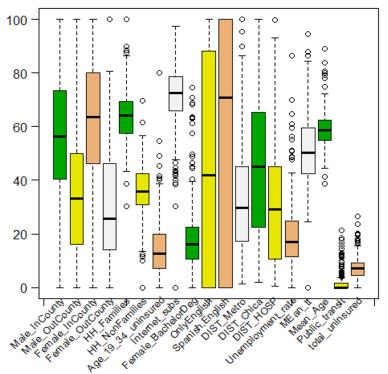
| Dimension | Variables | Categories: Values are in %. |
|------------------|---------------------|---|
| Demo- graphic | Age | Mean Age Under 19, 20 - 29, 30 - 44, 45 - 59, Above 60. |
| | Household Income | Mean Income Less than 10000, 10000 – 15000, 15000 – 25000, 25000 – 35000, 35000 – 50000, 50000 – 75000, 75000 – 100000, 100000 – 150000, 150000 – 200000, More than 200000. |
| | Race | Whites, African American, American Indian, Asian, Other races. |
| | Education | Male: High School Diploma, Male: Bachelor's or Higher Deg, Female: High School Diploma, Female: Bachelor's or Higher Deg. |

| Dimension | Variables | Categories: Values are in %. |
|----------------|--|---|
| Technology | Internet Subscription | Household with Internet subscription. |
| | Travel time | Mean commuting time to work |
| Transportation | Mode of transportation to work | Taxi, Public Transit |
| | Fertility | Women gave birth in the last 12 months |
| Others | Recently moved to other places | Male: within county, Male: to a different county, Female: within county, Female: to a different county. |
| Distance | Distance to Chicago To the nearest Hospital Distance to Metropolitan | Distance values are standardized by its maximum value |

- Remove variables that do not vary significantly across cities if Inter-Quartile Range (IQR) < 10.
- Result in 16 variables to which we added 3 that we thought were significant.



 Result in 16 variables to which we added 3 that we thought were significant.

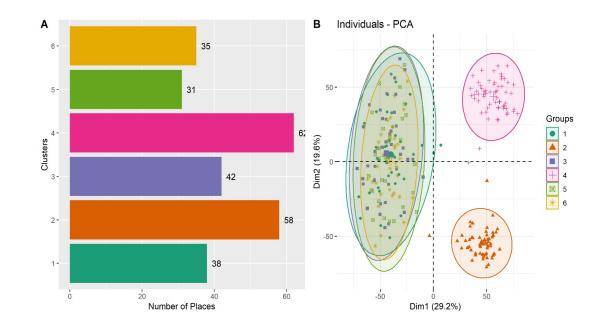


| Dimension | Variables | Categories: Values are in %. |
|-----------|------------------------|--|
| | Language | English only Spanish and English |
| | Household Structure | Families, Non-Families. |
| | Uninsured | Age: 19 - 34, |
| Social | | Uninsured percentage |
| | Employment | Male: Work in same county, Male: Work in different county, |
| | | Female: Work in same county, |
| | | Female: Work in different county. |
| | Unemployment | Unemployment percentage |

| Dimension | Variables | Categories: Values are in %. |
|------------------|---------------------|--------------------------------------|
| Demo- graphic | Age | Mean Age |
| | Household Income | Mean Income |
| | Education | Female: Bachelor's or Higher Deg. |

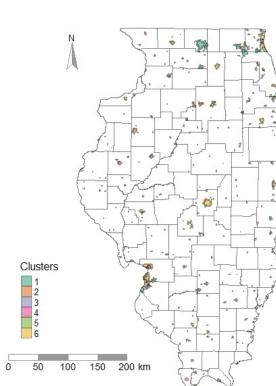
| Dimension | Variables | Categories: Values are in %. |
|------------|---|---|
| Technology | Internet Subscription | Household with Internet subscription. |
| | Mode of transportation to work | Public Transit |
| Distance | Distance to Chicago To the neares Hospital Distance to Metropolitan | – Distance values are † standardized by its – maximum value |

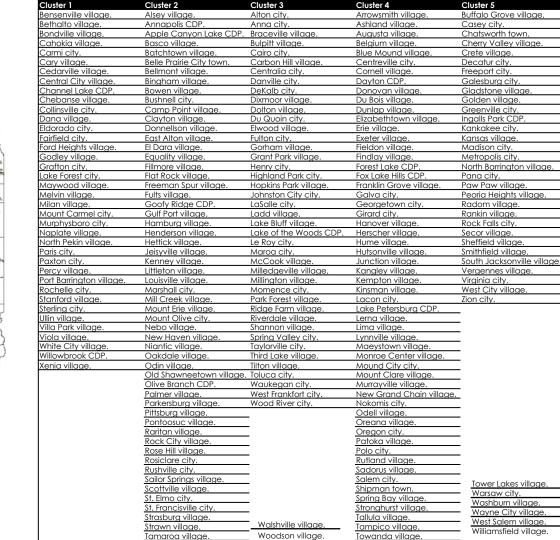
- Fuzzy C-Means (FCM) clustering.
- Evaluated with silhouette score.
- Plotted results using PCA.
- Found six clusters.



| Clusters | Variables | Characteristics |
|-----------|---|--|
| Cluster 1 | Unemployment rate | Around 24 % of the cities has UR exceeds 30%. |
| | Internet subscription | Cities in this cluster have highest mean average internet subscription. |
| | Distance to Hospitals | More cities in this cluster have hospitals nearby. |
| | Average age of population | 56% of the cities in this cluster have population with age above 60 . Meaning, this cluster has relatively higher older population. |
| Cluster 2 | Average Household Income | Cities with lowest mean household income. |
| | Distance to Chicago | Average distance to Chicago is higher for cities in this cluster. |
| Cluster 3 | Average commuting time | 21.4% of cities in this cluster have average commuting time of 65 minutes or longer . |
| | Distance to Metropolitan areas | Cities that are closest to a metropolitan area. |
| | Minorities | Has the highest average percentage of minorities living in these cities. |
| | Distance to Chicago | Cities in this cluster are relatively closer to Chicago . |
| Cluster 4 | Language | Cities in this cluster has, on average, the highest percentage of people who speak Spanish alone (52%). |
| | Income bracket: 75 to 100K with average percentage of population in this bracket is 12% | Highest number of cities (24%) in this cluster have more than 18% of population in this income bracket. |
| | Distance to Hospital | Cities in this cluster, on average, have longer distance to hospitals. |
| | Average commuting time | Highest number of cities (25%) with commuting time less than 40 minutes. |
| Cluster 5 | Family | This cluster, on average, has the highest percentage of people who does not live with family. |
| | Age | On average, this cluster has cities with highest percentage of people aged between 20 and 29 . |
| Cluster 6 | Education | Highest number of cities (20%) that have more than 30% of female with bachelor's degree. |
| | Transportation means to work | On average, this cluster has cities with highest percentage of people who walked to work . |

Clusters





Cluster 6

Armington village

Beaverville village.

Carbon Cliff village.

Bement village.

Canton city.

Cisne village.

Dupo village.

Dix village

Crystal Lake city.

Fast St. Louis city

Granite City city.

Granville village

Hodakins village.

Island Lake village

Lawrenceville city.

Mount Carroll city.

Mount Morris village

North Chicago city.

Pleasant Hill village.

Riverwoods village.

Sleepy Hollow village.

Washinaton Park village

Hoopeston city.

Jacksonville city

Macomb city.

Ottawa city.

Perry village

Pierron village.

Rockford city.

Steger village.

Villa Grove city.

Ursa village.

Peotone village.

Forest Park village

Barrinaton Hills village.

- Survey developed.
- Initial screening completed.
- Coding survey into Qualtrics completed.
- Pilot survey (in progress)
- Seek IRB exemption (in progress)

- Questions related to mobility challenges and opportunities:
 - Access to car
 - Access to transit and coordination between agencies
 - Presence of TNCs (e.g., Uber, Lyft)
 - Main challenges by mode (e.g., fuel costs, service frequency)
 - Solution identification (e.g., road maintenance, extended service)
 - Impact on people with physical disabilities
 - ...



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Impact of COVID-19

COVID-19 Future Survey

The 'COVID Future Panel Survey' is nationwide online panel survey that collects the information on attitudinal and behavioral changes before, during, and after the pandemic.

The survey asked questions about commuting, daily travel, air travel, working from home, online learning, shopping, and risk perception, along with attitudinal, socioeconomic, and demographic information.

The survey data was properly weighed and cleaned and shared publicly on ASU Dataverse.

The project was funded by the NSF RAPID Award 2030156.

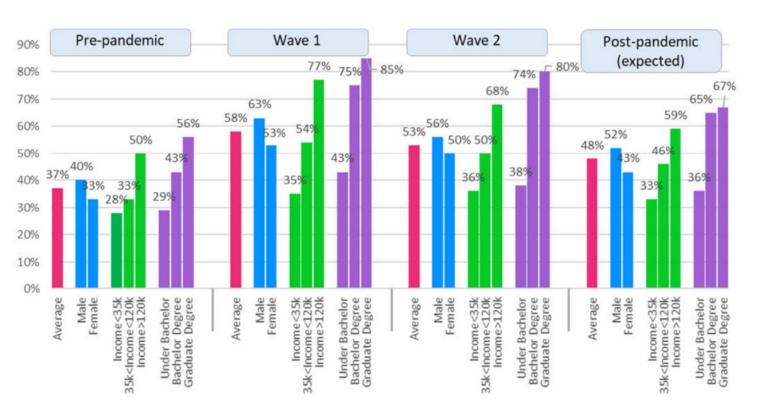
COVID-19 Future Survey

The survey was conducted in multiple waves, by reaching out to the same respondents over time.

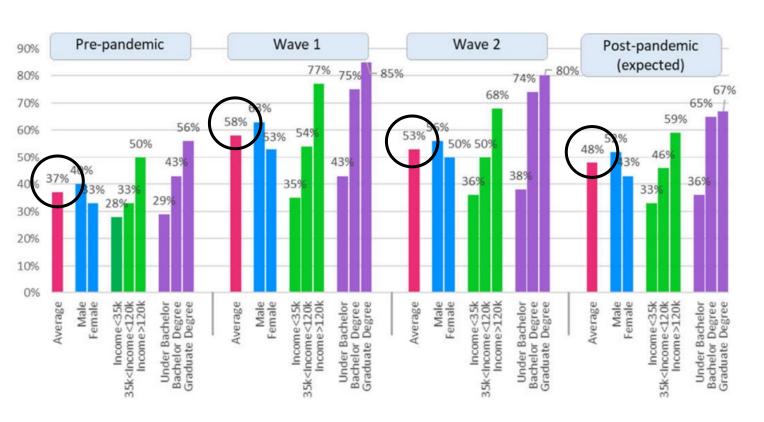
The Wave 1 of the survey was conducted from April 2020 to October 2020; Wave 2 from November 2020 to May 2021; and Wave 3 started in October 2021 and is ongoing.

The total responses in wave 1 were 8,723 and that in wave 2 were 2,973.

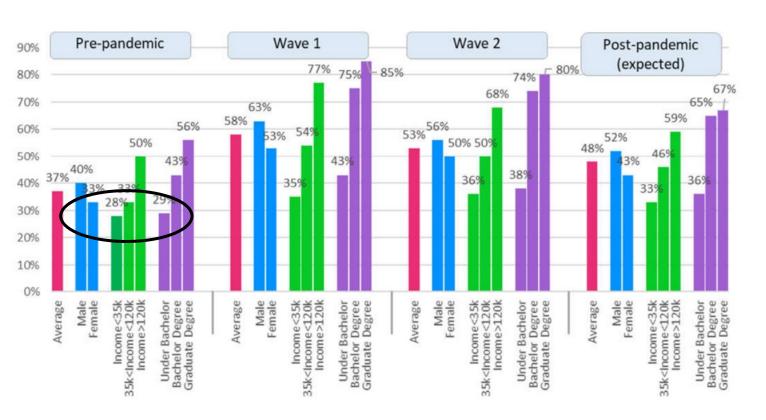
For more information about the survey, please visit <u>covidfuture.org</u>.



Workers who have not been laid off during the pandemic were asked if they have the option to telecommute?

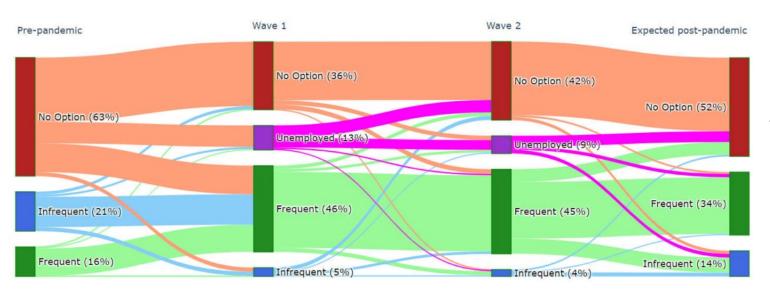


Through the pandemic, there has been an increase in the percentage of respondents who have the option to telecommute. A large percentage of people also expect to be able to telecommute after the pandemic.

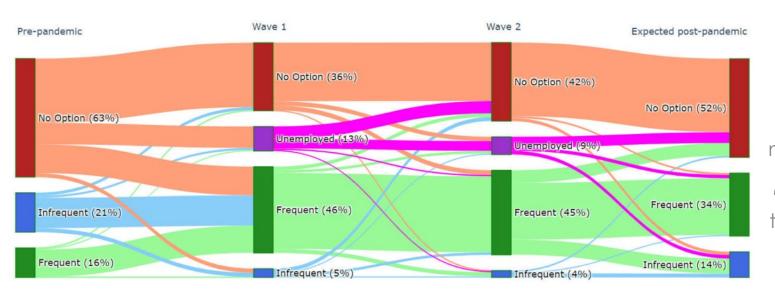


Following have below average proportion with option to telecommute:

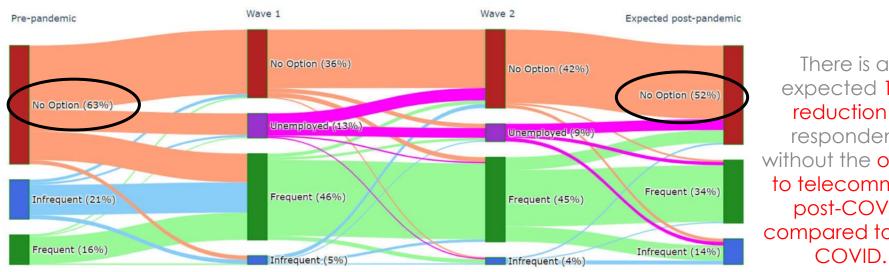
- Females
- Household income less than 120 K
- Education less than Bachelor's degree.



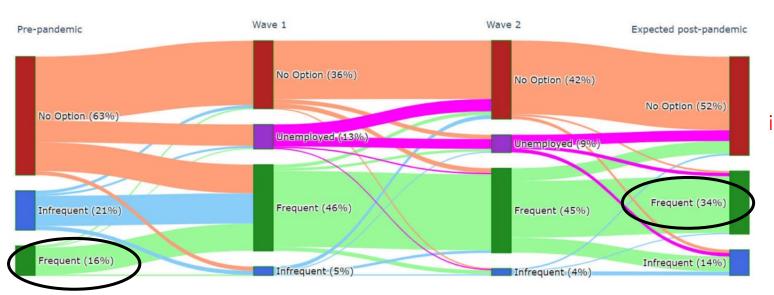
Next, we look at the telecommute frequency before, during and (expected) after COVID-19.



Here 'frequent'
refers to
telecommuting
more than once a
week. Whereas,
'infrequent' refers
to telecommuting
once a week or
less.



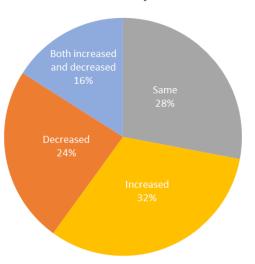
There is an expected 17% reduction in respondents without the option to telecommute post-COVID compared to pre-



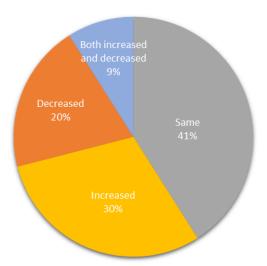
Frequent commuters' increase from 16% pre-pandemic to 34% in postpandemic (i.e., 112% growth).

Work Productivity

Wave 1 vs Pre-pandemic

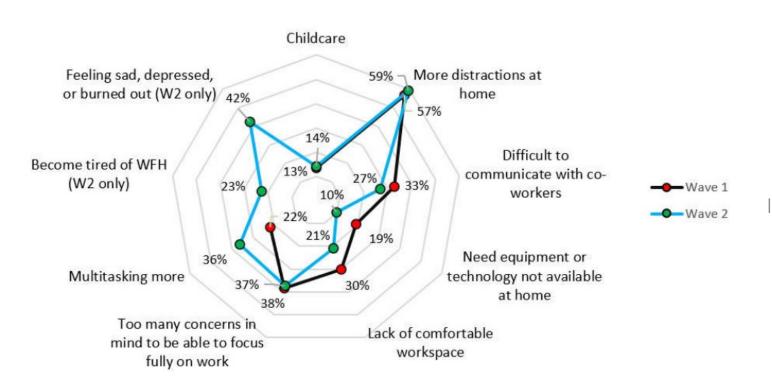


Wave 2 vs Pre-pandemic



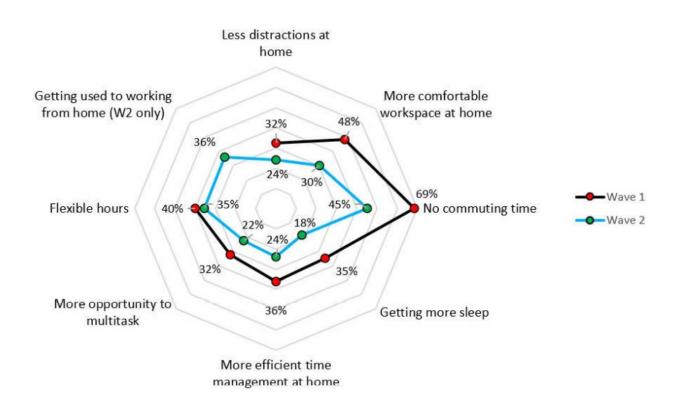
60% of respondents reported that there work productivity increased or remained same in wave 1 (earlier in pandemic). While this percentage increased to 71% in wave 2 (later in pandemic).

Decrease Work Productivity



More distractions at home is the biggest factor in both waves responsible for lower productivity.

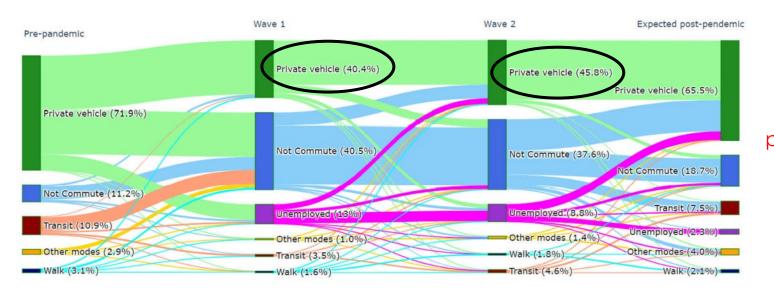
Increase Work Productivity



No commuting time is the biggest factor in both waves responsible for higher productivity.

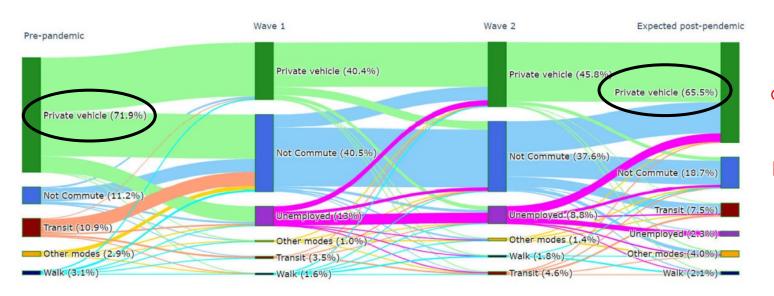
Commuting

Commuting Mode Choice



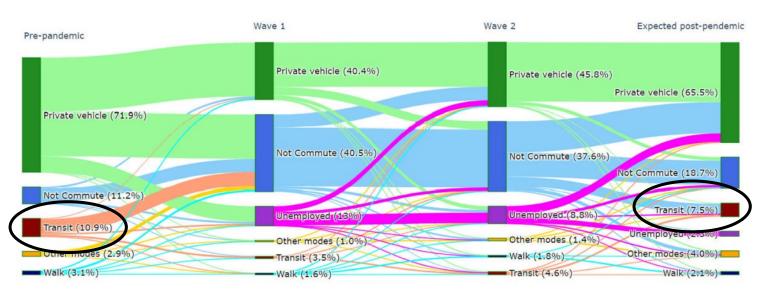
The share of private vehicle commuters plummeted to 40% in wave 1 and remained at around 45% in wave 2.

Commuting Mode Choice



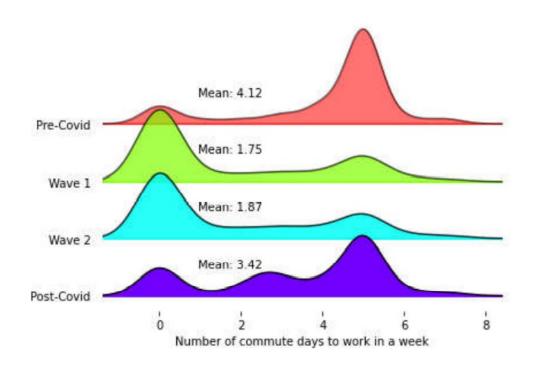
Post-pandemic, only about 66% of the respondents expect to use a private vehicle to commute. While around 19% expect not to commute.

Commuting Mode Choice



Transit share kept increasing since wave 1 but is still expected to be significantly lower than the prepandemic.

Commute Frequency



Pre-pandemic, the average number of commute days were 4.1 days/week, which reduced to 1.75 days/week in wave 1 and 1.87 days/week in wave 2. Post-pandemic, it is expected get to 3.42 days/week.

Mode Use Expectation for Post-COVID



Next, we looked at mode use (for all purposes, not just commuting) postpandemic

Expected Post-COVID Use

Mode Use Expectation for Post-COVID



Expected Post-COVID Use

12% expect to use private vehicles less than before, whereas 17% expect to use them more.

While the proportion of private vehicle commute trips are expected to decrease, the overall use of private vehicles might increase.

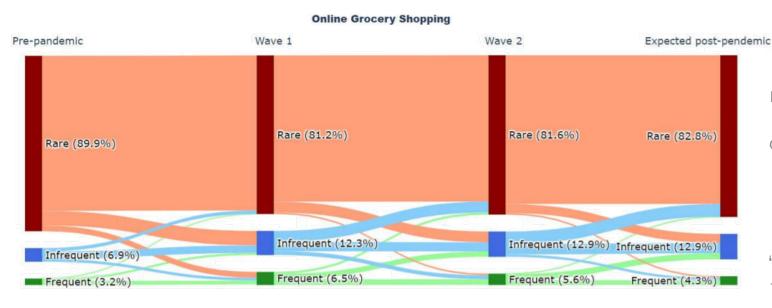
Mode Use Expectation for Post-COVID



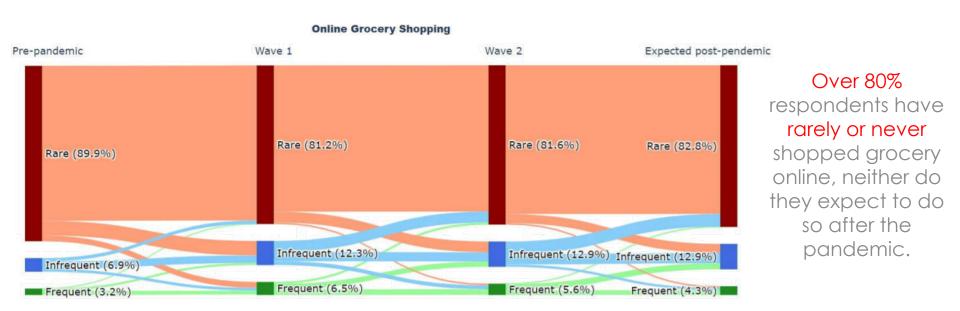
13% expect to use transit less than before, while 10% expect to use transit more than before.

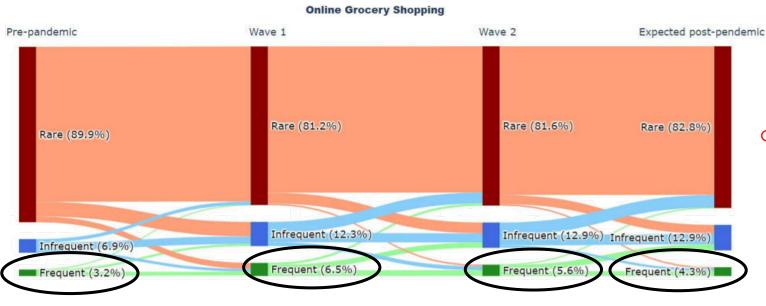
Expected Post-COVID Use

Online Shopping

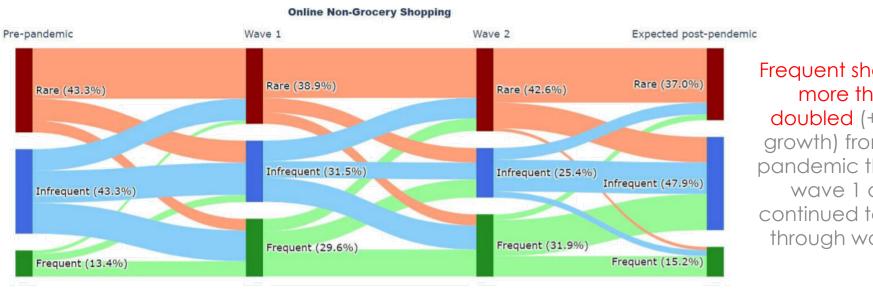


Here 'frequent'
refers to shopping
more than
once/week. While
'infrequent' refers
to that between
once/week and
once/month).
'Rare' refers to less
than once/month
or never.





Frequent online grocery shoppers have roughly doubled in wave 1 but have slightly decreased in wave 2. A higher percentage of people expect to grocery shop online postpandemic.



Frequent shoppers more than doubled (+121% growth) from prepandemic through wave 1 and continued to grow through wave 2.

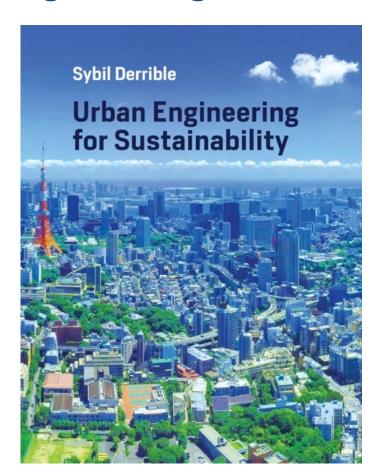
Integration with other infrastructure

Urban Infrastructure



| | Transport | Water | Utility | Electricity | Telecom | Solid Waste | Buildings |
|----------------|---|--|--|--|---|--|---|
| Transport | | Underground water conduits in streets Leaks and runoff leading to street flooding Overflowing of stormwater channels leading to flooding | Underground utility lines in streets Occasional construction and maintenance of infrastructure leading to traffic disruption | Raw material transport for electricity generation Electricity needed for electricity vehicles, electric rail and bus modes and for operations (e.g., traffic signals, street lights) | Underground telecom lines in streets Transmission of real- time information | Bins / cans located on sidewalks, back alleys, roads, etc. Solid waste collection and transfer vehicles use roads. Land reclamation create space for transport infrastructure. | Conflict for land Buildings as location where people go to or depart from |
| Water | Restricted right-of-way Hard to reach water infrastructure when located underground Impermeable surfaces leading to flooding | | Competition for underground space Gas-run pumps for water distribution Gas leak can contaminate groundwater wells | Competition for underground space Electricity to treat and distribute water (Energy-Water Nexus) | Competition for underground space Information to manage water distribution systems (e.g., SCADA) Increasing reliance on telecom with smart meters | Contamination of surface water bodies and aquifers with incineration and landfilling. Ability of waste facilities to receive solid waste from treatment plants. | Force water conduits to be below streets Impermeable surfaces leading to flooding Buildings as places of water consumption |
| Utility | Restricted right-of-way Hard to reach gas lines as well as steam and chilled water pipes when located underground | Competition for underground space | | Competition for underground space | Competition for underground space Information transmission for real time monitoring | Predictable generation of methane for natural gas and district heating systems. | Buildings as places of gas consumption Buildings as places of steam and chilled water consumption for space heating |
| Electricity | Restricted right-of-way Hard to reach distribution infrastructure when located underground Movement of raw material for electricity generation | Competition for underground space Thermal power systems require significant amounts of water (Energy-Water Nexus) | Competition for underground space Electricity generation from natural gas | | Competition for underground space Similar to water, increasing reliance on telecom with smart meters | Predictable generation of electricity. Ability of waste facilities to receive solid waste from power plants (e.g., nuclear waste). | Partially directs how distribution lines are installed Buildings as places of electricity consumption Hazard with tree branches next to buildings |
| Telecom | Restricted right-of-way Hard to reach telecom lines when located underground Many Internet cables are located next to rail tracks | Competition for underground space Large amounts of water are needed for cooling, especially in data centers | Competition for underground space | Competition for underground space All telecom devices require electricity Data centers require a significant amount electricity for cooling | | Ability of waste facilities to receive solid waste from telecom (e.g., wires). | Buildings as end points where telecom lines are installed |
| Solid Waste | Roads must be accessible for solid waste collection and transport vehicles. Space must be dedicated to solid waste infrastructure | Some processes require stable supply of water. Heavy rains to impact landfilling activities. Facilities use water. | Natural gas needed to initiate / aid combustion. Heating / cooling solid waste facilities. | Some processes require stable supply of electricity (e.g., eddy current separators). Facilities use electricity. | Environmental monitoring of landfills. Increasing reliance on telecom (e.g., GPS in garbage trucks) | | Solid waste generated in buildings. Periodic service of solid waste collection. Buildings host solid waste facilities. |
| Buildings | Conflict for land Building location (e.g., in real estate) | Presence / availability of water Water problems lead to flooding (e.g., basement) Conflict for land for larger water infrastructure | Presence / availability of gas Systems' size for district heating/cooling Conflict for land for larger gas infrastructure | Presence / availability of electricity Conflict for land for larger transmission lines | Presence / availability of telecom lines Buildings are sometimes strategically located to be near a main telecom hub | Accommodating solid waste generation (e.g., trash chute, dumpster at back). Ability of waste facilities to receive solid waste from buildings. | 51 |

Urban Engineering for Sustainability



Sybil Derrible

Urban Engineering for Sustainability

MIT Press 2019

Opportunities?

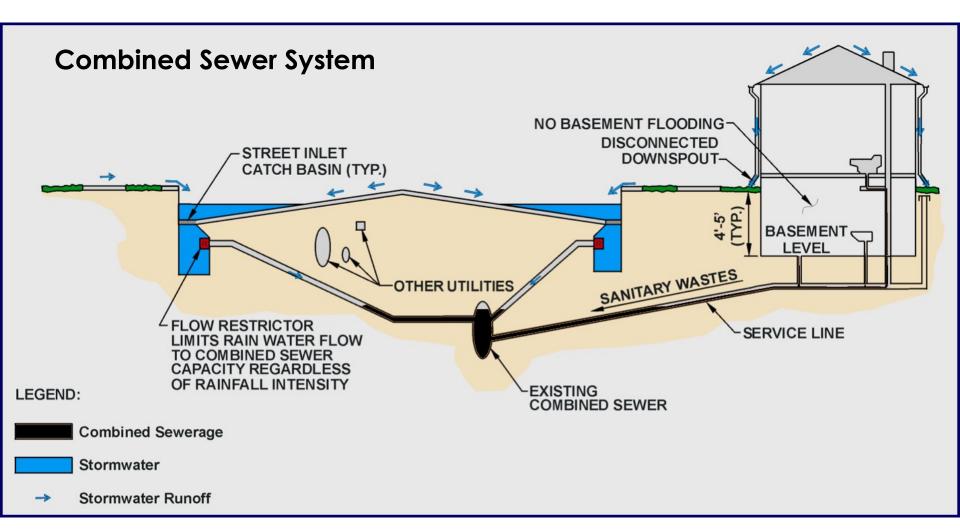
- Lower energy consumption
- Better monitor road conditions
- Provide better access to fiberoptic Internet
- Improve stormwater management

Transforming Cities through Transportation & Stormwater Management

Wastewater

Two Types

- 1. Sanitary
- 2. Stormwater



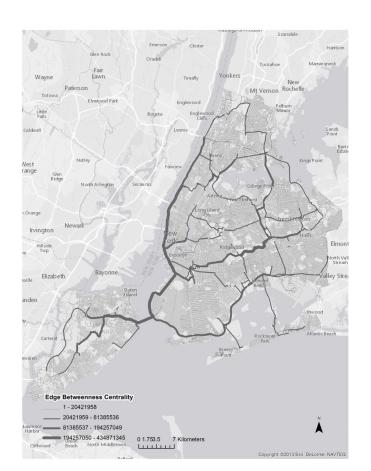
Separate Sewer System

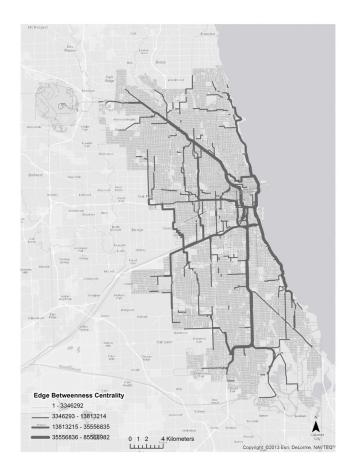




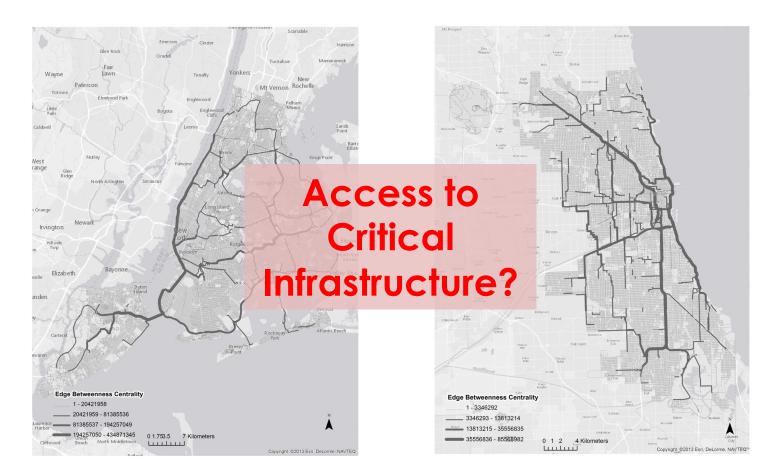
Road Resilience

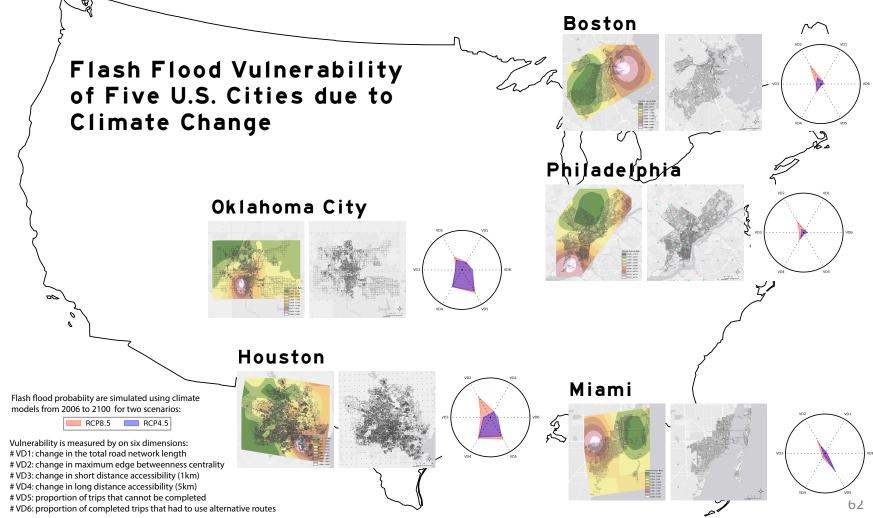
Road Resilience – Flooding





Road Resilience - Flooding





Conclusion

- Many cities in Illinois at depopulating. Join our study!
- COVID-19 is transforming how we live. In particular, more people will more frequently work from home. Transit will need to adapt.
- Transportation is integrated with other infrastructure systems.
 Make these connections count!

Thank You

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Acknowledgements





